

Lecture 3. Discrete Random Variables

January 23, 2013

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- Cumulative Distribution Functions
- Properties of CDFs
- Discrete Random Variables
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 - ▶ The Bernoulli Distribution
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 - ▶ The Geometric Distribution
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Motivation and Definition

Statistics is concerned with **data**.

Question: How do we link **sample spaces** and **events** to **data**?

Answer: The link is provided by the concept of a **random variable**.

Definition

A random variable is a mapping $X : \Omega \rightarrow \mathbb{R}$ that assigns a real number $x = X(\omega)$ to each realization $\omega \in \Omega$.

Remark: Technically, a random variable must be a **measurable function**.

Example: Flip a coin 10 times. Let $X(\omega)$ be the number of heads in the sequence. For example, if $\omega = HHTHTTTHTH$, then $X(\omega) = 5$.

Given a **random variable** X and a set $A \subset \mathbb{R}$, define

$$X^{-1}(A) = \{\omega \in \Omega : X(\omega) \in A\}$$

and let

$$\begin{aligned}\mathbb{P}(X \in A) &= \mathbb{P}(X^{-1}(A)) = \mathbb{P}(\{\omega \in \Omega : X(\omega) \in A\}) \\ \mathbb{P}(X = x) &= \mathbb{P}(X^{-1}(x)) = \mathbb{P}(\{\omega \in \Omega : X(\omega) = x\})\end{aligned}$$

The Cumulative Distribution Function

Definition

The cumulative distribution function (CDF) $F_X : \mathbb{R} \rightarrow [0, 1]$ is defined by

$$F_X(x) = \mathbb{P}(X \leq x)$$

Example: Flip a fair coin twice and let X be the number of heads.

Find the CDF of X

Question: Why do we bother to define CDF?

Answer: CDF effectively contains all the information about the random variable

Theorem

Let X have CDF F and Y have CDF G . If $F(x) = G(x)$ for all x , then $\mathbb{P}(X \in A) = \mathbb{P}(Y \in A)$. In words, the CDF completely determines the distribution of a random variable.

Properties of CDFs

Question: Given a function $F(x)$, can we find a **random variable** X such that $F(x)$ is the CDF of X , $F_X(x) = F(x)$?

Theorem

A function $F : \mathbb{R} \rightarrow [0, 1]$ is a CDF for some random variable if and only if it satisfies the following three conditions:

① F is **non-decreasing**:

$$x_1 < x_2 \Rightarrow F(x_1) \leq F(x_2)$$

② F is **normalized**:

$$\lim_{x \rightarrow -\infty} F(x) = 0 \quad \text{and} \quad \lim_{x \rightarrow +\infty} F(x) = 1$$

③ F is **right-continuous**:

$$\lim_{y \rightarrow x+0} F(y) = F(x)$$

Discrete Random Variables

Definition

X is **discrete** if it takes countable many values $\{x_1, x_2, \dots\}$.
We define the **probability mass function** (PMF) for X by

$$f_X(x) = \mathbb{P}(X = x)$$

Example: Flip a fair coin twice and let X be the number of heads.
Find the probability mass function of X .

The **CDF** of X is related to the **PMF** f_X by

$$F_X(x) = \mathbb{P}(X \leq x) = \sum_{x_i \leq x} f_X(x_i)$$

The **PMF** f_X is related to the **CDF** F_X by

$$f_X(x) = F_X(x) - F_X(x^-) = F_X(x) - \lim_{y \rightarrow x-0} F(y)$$

Important Examples

- **The Point Mass Distribution**

X has a **point mass** distribution at a , denoted $X \sim \delta_a$, if $\mathbb{P}(X = a) = 1$.
In this case

$$F(x) = \begin{cases} 0, & x < a; \\ 1, & x \geq a. \end{cases}$$

and

$$f(x) = \begin{cases} 1, & x = a; \\ 0, & x \neq a. \end{cases}$$

- **The Discrete Uniform Distribution**

Let $n > 1$ be a **given integer**. Suppose that X has probability mass function given by

$$f(x) = \begin{cases} 1/n, & \text{for } x = 1, \dots, n; \\ 0, & \text{otherwise.} \end{cases}$$

We say that X has a uniform distribution on $1, \dots, n$.

Important Examples

- **The Bernoulli Distribution**

Let X represents a **coin flip**. Then $\mathbb{P}(X = 1) = p$ and $\mathbb{P}(X = 0) = 1 - p$ for some $p \in [0, 1]$. We say that X has a Bernoulli distribution, denoted $X \sim \text{Bernoulli}(p)$. The probability mass function is

$$f(x|p) = p^x(1 - p)^{1-x}, \quad x \in \{0, 1\}$$

- **The Binomial Distribution**

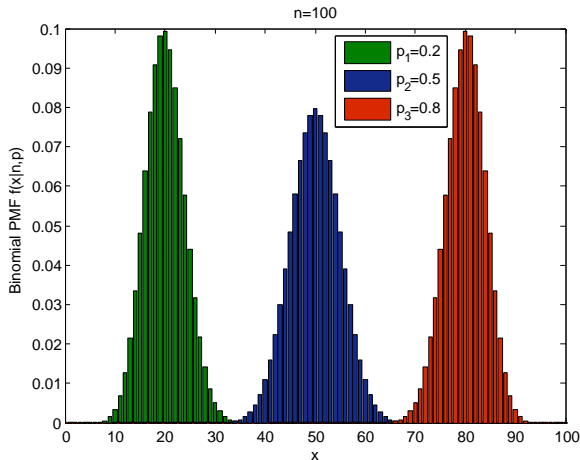
Suppose we have a coin which falls heads with probability p for some $p \in [0, 1]$. Flip the coin n times and let X be the **number of heads**. Assume that the tosses are independent. The probability mass function of X is then

$$f(x|n, p) = \begin{cases} \binom{n}{x} p^x (1 - p)^{n-x}, & \text{if } x = 0, 1, \dots, n; \\ 0, & \text{otherwise.} \end{cases}$$

A random variable with this mass function is called a Binomial random variable and we write $X \sim \text{Bin}(n, p)$.

Remark: X is a **random variable**, x denotes a **particular value** of the random variable, n and p are **parameters**, that is, fixed real numbers. The parameter p is usually **unknown** and must be estimated from **data**.

Binomial Distribution $\text{Bin}(n, p)$



Important Examples

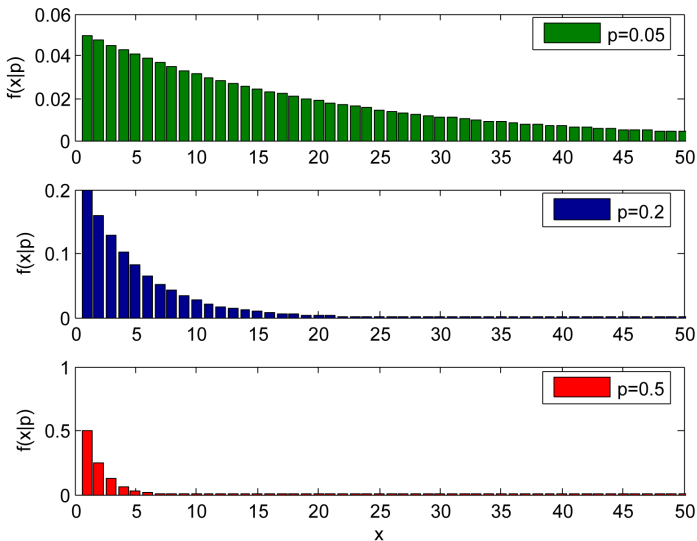
- **The Geometric Distribution**

X has a geometric distribution with parameter $p \in (0, 1)$, denoted $X \sim \text{Geom}(p)$, if

$$f(x|p) = p(1 - p)^{x-1}, \quad x = 1, 2, 3 \dots$$

Think of X as the **number of flips needed until the first heads** when flipping a coin. Geometric distribution is used for modeling the **number of trials until the first success**.

Geometric Distribution $\text{Geom}(p)$



Important Examples

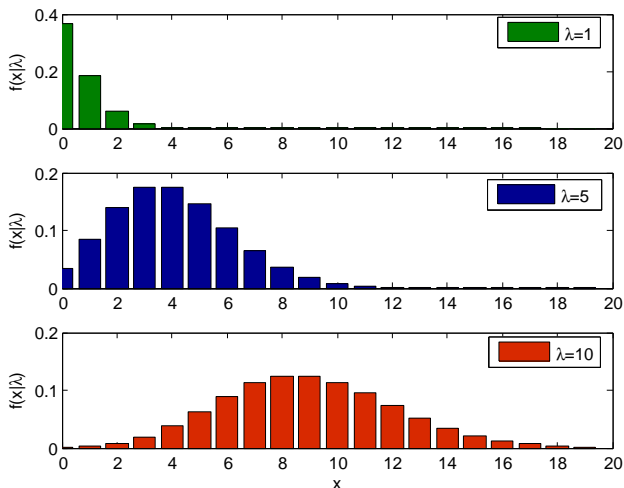
- **The Poisson Distribution**

X has a Poisson distribution with parameter λ , denoted $X \sim \text{Poisson}(\lambda)$ if

$$f(x|\lambda) = e^{-\lambda} \frac{\lambda^x}{x!}, \quad x = 0, 1, 2, \dots$$

The Poisson distribution is often used as a model for counts of **rare events** like traffic accidents. $f(x|\lambda)$ expresses the probability of a given **number of events** x occurring in a fixed interval of time if these events occur with a known **average rate** λ and **independently** of the time since the last event.

Poisson Distribution $\text{Poisson}(\lambda)$



Summary

- A **random variable** is a mapping $X : \Omega \rightarrow \mathbb{R}$ that assigns a real number $x = X(\omega)$ to each realization $\omega \in \Omega$.
- The **cumulative distribution function** (CDF) is defined by

$$F_X(x) = \mathbb{P}(X \leq x)$$

- ▶ CDF **completely determines** the distribution of a **random variable**
 - ▶ CDF is **non-decreasing**, **normalized**, and **right-continuous**
- Random variable X is **discrete** if it takes **countable many values** $\{x_1, x_2, \dots\}$.
- The **probability mass function** (PMF) of X is

$$f_X(x) = \mathbb{P}(X = x)$$

- Relationships between **CDF** and **PMF**:

$$F_X(x) = \mathbb{P}(X \leq x) = \sum_{x_i \leq x} f_X(x_i)$$

$$f_X(x) = F_X(x) - F_X(x^-)$$